RELAMPAGO LMA Level 3 File Documentation

Level 3 data were created using the Imatools Python package (https://github.com/deeplycloudy/Imatools). See references below for more information about LMA data, processing, and each variable.

Each RELAMPAGO LMA Level 3 file is a netCDF4 architecture. Note that RELAMPAGO LMA data always start (and end) 1 second after each 10-minute period (e.g., 000001, 001001, 002001, etc.). This is an artifact of how the Level 0 data were recorded.

All data were processed with a chisqr maximum of 5, and requiring at least 5 VHF sources to identify a flash. Data are gridded to 1-km resolution grids (all dimensions) and binned every minute. New files are available every 10 minutes. Two-dimensional data structures are (ntimes, lon, lat) and 3D data structures are (ntimes, lon, lat, alt). Number of bins in each dimension are $10 \text{ (ntimes)} \times 400 \text{ (lon)} \times 401 \text{ (lat)} \times 20 \text{ (alt)}$

Info about files

- *_flash_extent.nc & *_flash_extent_3d.nc: Flash extent density (km⁻² or km⁻³; i.e., number of flashes that pass through a given pixel/cube in the given time bin)
- *_flash_init.nc & *_flash_init_3d.nc: Number of flashes that initiated within the given pixel/cube and time bin
- *_footprint.nc & *_footprint_3d.nc: Footprints (mean area/volume in km²/km³) of flashes that occurred within pixel/cube and time bin
- *_specific_energy.nc & *_specific_energy _3d.nc: Specific energy (mean; J/kg) of flashes that occurred within pixel/cube and time bin
- *_total_energy.nc & *_total_energy _3d.nc: Total energy (J) of flashes that occurred within pixel/cube and time bin
- *_source.nc & *_source_3d.nc: Number of sources (km⁻² or km⁻³) that occurred in grid element
- *_flashsize_std.nc & *_flashsize_std_3d.nc: Standard deviation (km² or km³) of flash sizes within grid element

Each file also contains dimensional variables such as time, longitude, latitude, and altitude.

In Python, you can retrieve level 3 data into numpy arrays using the following example code snippet:

```
import numpy as np
import xarray
data = xarray.open_dataset('REL_20181214_230001_600_5src_0.0105deg-
dx_flash_extent.nc')
lon = np.array(data.lon.data)
lat = np.array(data.lat.data)
bintime = np.array(data.time.data)
flash_extent = np.array(data.flash_extent.data)
data.close()
```

References

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Fuchs, B. R., Bruning, E. C., Rutledge, S. A., Carey, L. D., Krehbiel, P. R., & Rison, W. (2016). Climatological analyses of LMA data with an open-source lightning clustering algorithm. Journal of Geophysical Research: Atmospheres, 121, 8625-8648. https://doi.org/10.1002/2015JD024663

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